

A Data-Driven Approach in Designing RST Controllers with H_∞ performance via Convex Optimization

Achille Nicoletti, Zlatko Emedi and Alireza Karimi

Abstract—In this paper, a new method for designing robust fixed-order \mathcal{H}_∞ discrete-time controllers is presented. The controller structure is a two-degree of freedom polynomial controller of the RST-type. A data-driven approach is implemented for the design process in order to capture the unmodeled dynamics that may exist with parametric models. The \mathcal{H}_∞ robust performance condition can be represented by a set of convex constraints with respect to the parameters of the RST controllers. A convex optimization algorithm can then be implemented to obtain these parameters. The proposed method is applied to a multi-axis torsional system where the goal is to control the position of the disks with variable inertias.

I. INTRODUCTION

In many of today's complex industrial applications, it is difficult to model a process with extreme precision. For such processes, a high-order model is typically required to capture the dynamics of a system; however, high order models lead to high order controllers with numerical implementation problems. Therefore, in practice, low-order models are preferred in order to simplify the controller design process. However, low-order models possess unmodeled dynamics that can hinder the performance of a controller. Robust controller design methods use the uncertainty models as weighting filters to ensure the stability and performance of the closed-loop system. This in turn increases the controller order which is equal to the order of the plant model and the uncertainty weighting filter. In data-driven methods, the plant model is represented by a set of data; thus there are no unmodeled dynamics and the controller order can be chosen independently of the plant model complexity. The data-driven approach can be realized in two manners: using time-domain or frequency-domain data. In this paper, the frequency-domain approach will be utilized for the controller design scheme.

The field of frequency-domain controller design techniques continues to spark the interest of many researchers. In [1], a controller is designed in the frequency-domain by considering a least-squares optimization approach; however, in this method, it is not evident on how to predefine the structure of a stabilizing controller. A data-driven design method that guarantees the stability of a closed-loop system by setting gain and phase margin constraints in the Nyquist diagram has been addressed in [2]. Despite the simplicity and advantages of this method, it cannot be applied to unstable plants. Moreover, the constraints are conservative and may lead to a controller that inhibits the system performance.

A robust frequency-domain control design method has been established in [3]; this method, however, requires a solution to a non-linear optimization problem.

Robust control methods are continually addressed within the control system community. Many research papers and books have been devoted to developing new algorithms in designing robust controllers that reduce the conservatism associated with uncertainty modeling. Robust controller design methods belonging to the \mathcal{H}_∞ control framework minimize the \mathcal{H}_∞ norm of a weighted closed-loop sensitivity function. An \mathcal{H}_∞ loop-shaping method to design stabilizing controllers has been examined in [4]. However, a non-convex optimization algorithm is required to obtain a solution. Generally, it is desired to acquire a convex optimization problem since convex problems are computationally tractable. A convex approximation of the \mathcal{H}_∞ criterion has been discussed in [5] and [6] where convex constraints are imposed on the Nyquist diagram. The controller is linearly parameterized (the denominator is fixed) and a desired open loop transfer function is used to convexify the \mathcal{H}_∞ constraints. An extension to this method has been presented in [7], where a single-input-single-output (SISO) controller is represented in a rational form (and thus allowing the numerator and denominator of a controller to be optimized separately). This method gives the necessary and sufficient conditions for the existence of a robust controller for systems represented by a frequency response function.

The RST controller structure is an effective discrete-time two-degree of freedom polynomial controller where the tracking and regulation characteristics of a closed-loop system can be formulated independently [8]. Various methods and applications of the RST controller design methodology have been addressed in [9], [10], [11], and [12]. However, in these previous applications, the controllers were devised based on the knowledge of the dynamic model of the physical system. A frequency domain approach in designing the RST controllers has been discussed in [13]. In this approach, a convex optimization algorithm was formulated by considering a convex approximation of the \mathcal{H}_∞ criterion. However, in order to preserve convexity, this method required one of the controller terms (R) to be fixed a priori.

The proposed method in this paper is an extension of [13] and [7], where an RST controller will be designed by formulating a convex optimization problem in the frequency domain. In [13], the controller R was fixed with an integrator¹, while S and T were linearly parameterized. However,

Automatic Control Laboratory, Ecole Polytechnique Fédérale de Lausanne (EPFL), 1015 Lausanne, Switzerland

Corresponding author: alireza.karimi@epfl.ch

¹Note that the variable names for R and S are interchanged in [13]

with the methodology presented in this paper, the R , S and T controllers can each be linearly parameterized, and thus introduces more degrees of freedom which improves the \mathcal{H}_∞ performance. Moreover, new convex constraints are imposed in order to guarantee that the closed-loop system remains stable while attaining the desired tracking performance without the need to specify a desired open-loop transfer function.

This paper is organized as follows: In section (II), the class of models, controllers and control objectives are defined. Section (III) will discuss the control design methodology and stability conditions of the proposed method for the RST controller structure. Convex conditions will be formulated based on the \mathcal{H}_∞ criterion. Section (IV) will demonstrate the effectiveness of the method by applying the proposed design scheme to a multi-axis torsional system. Finally the concluding remarks are given in Section (V).

Notation: In order to avoid the risk of any confusion, the notation for the symbols employed in this paper will be defined here.

\mathbb{R} : the set of all real numbers.

\mathbb{Z} : the set of all positive integers including zero.

$\text{Re}\{\cdot\}$: the real part of a complex variable.

$\text{Im}\{\cdot\}$: the imaginary part of a complex variable.

a^\top : the transpose of the vector a .

z : variable used to represent the complex frequency domain of discrete-time systems.

s : variable used to represent the complex frequency domain of continuous-time systems.

II. PROBLEM FORMULATION

A. Class of models

Let us begin by considering the frequency response function (FRF) of a discrete-time SISO plant given as

$$G(e^{-j\omega}) = N(e^{-j\omega})M^{-1}(e^{-j\omega}), \quad \forall \omega \in \Omega \quad (1)$$

where $\Omega \in [0, \pi]$. $N(e^{-j\omega})$ and $M(e^{-j\omega})$ must be FRF's of bounded analytic functions outside the unit circle. For stable plants, a trivial choice is $N(e^{-j\omega}) = G(e^{-j\omega})$ and $M(e^{-j\omega}) = 1$. For unstable plants, $N(e^{-j\omega})$ and $M(e^{-j\omega})$ can be obtained, in a data-driven setting, by considering the acquired data of a closed-loop system with a stabilizing controller. If $N(e^{-j\omega})$ represents the FRF of the transfer function from the reference input to the output, and $M(e^{-j\omega})$ represents the FRF of the transfer function from the reference input to the controller output, then the FRF of the plant model can be obtained as asserted in (1).

In general, a set \mathcal{G} that represents a plant model containing k FRF models can be defined:

$$\mathcal{G} = \{G_i(e^{-j\omega}); \quad i = 1, \dots, k; \quad \forall \omega \in \Omega\} \quad (2)$$

These FRF's can be determined by considering the frequency response of a parametric model or from a set of input/output data. For simplicity, one model from the set \mathcal{G} will be considered, and the subscript i will be omitted. However, in general, the design procedures outlined in this paper can be

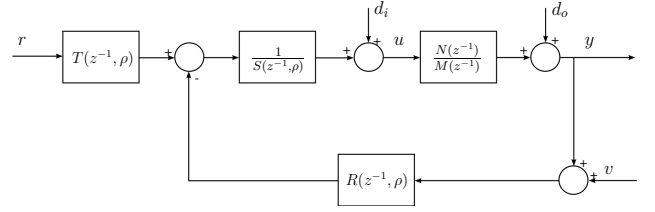


Fig. 1. RST controller structure

applied to the multi-model case. In section (IV), the proposed method is applied on a torsional apparatus with multi-model uncertainty.

B. Class of controllers

The controller structure that will be considered will be of the RST-type, as shown in Fig.1. The RST controllers are linearly parameterized in the variables r_k , s_k , and t_k ; they are each polynomials in z which can be expressed as:

$$R(z^{-1}) = r_0 + r_1 z^{-1} + \dots + r_{n_r} z^{-n_r} \quad (3)$$

$$S(z^{-1}) = 1 + s_1 z^{-1} + \dots + s_{n_s} z^{-n_s} \quad (4)$$

$$T(z^{-1}) = t_0 + t_1 z^{-1} + \dots + t_{n_t} z^{-n_t} \quad (5)$$

where $\{n_r, n_s, n_t\} \in \mathbb{Z}$. These controllers can also be represented in vector form as $R(z^{-1}, \rho) = \rho_R^\top \phi_R(z^{-1})$, $S(z^{-1}, \rho) = \rho_S^\top \phi_S(z^{-1})$ and $T(z^{-1}, \rho) = \rho_T^\top \phi_T(z^{-1})$, where

$$\rho_R^\top = [r_0, r_1, \dots, r_{n_r}]; \quad \phi_R^\top(z^{-1}) = [1, z^{-1}, \dots, z^{-n_r}] \quad (6)$$

$$\rho_S^\top = [1, s_1, \dots, s_{n_s}]; \quad \phi_S^\top(z^{-1}) = [1, z^{-1}, \dots, z^{-n_s}] \quad (7)$$

$$\rho_T^\top = [t_0, t_1, \dots, t_{n_t}]; \quad \phi_T^\top(z^{-1}) = [1, z^{-1}, \dots, z^{-n_t}] \quad (8)$$

C. Process Definitions

Since the design techniques introduced in this paper belong to the \mathcal{H}_∞ framework, it is appropriate to consider the various sensitivity functions associated with the controller structure shown in Fig. 1. Some of the sensitivity functions for this process can be asserted as follows:

$$\mathcal{S}_0(z^{-1}, \rho) = \frac{M(z^{-1})S(z^{-1}, \rho)}{\psi(z^{-1}, \rho)} \quad (9)$$

$$\mathcal{S}_i(z^{-1}, \rho) = \frac{N(z^{-1})S(z^{-1}, \rho)}{\psi(z^{-1}, \rho)} \quad (10)$$

$$\mathcal{T}(z^{-1}, \rho) = \frac{N(z^{-1})T(z^{-1}, \rho)}{\psi(z^{-1}, \rho)} \quad (11)$$

$$\mathcal{U}(z^{-1}, \rho) = \frac{M(z^{-1})T(z^{-1}, \rho)}{\psi(z^{-1}, \rho)} \quad (12)$$

$$\mathcal{V}(z^{-1}, \rho) = -\frac{N(z^{-1})R(z^{-1}, \rho)}{\psi(z^{-1}, \rho)} \quad (13)$$

$$\mathcal{E}(z^{-1}, \rho) = \frac{\psi(z^{-1}, \rho) - N(z^{-1})T(z^{-1}, \rho)}{\psi(z^{-1}, \rho)} \quad (14)$$

where $\psi(z^{-1}, \rho) = M(z^{-1})S(z^{-1}, \rho) + N(z^{-1})R(z^{-1}, \rho)$ and $\mathcal{S}_0(z^{-1}, \rho)$ is the transfer function between d_o and y , $\mathcal{S}_i(z^{-1}, \rho)$ is the transfer function between d_i and y ,

$\mathcal{T}(z^{-1}, \rho)$ is the transfer function between r and y , $\mathcal{U}(z^{-1}, \rho)$ is the transfer function between r and u , $\mathcal{V}(z^{-1}, \rho)$ is the transfer function between v and y , and $\mathcal{E}(z^{-1}, \rho)$ is the transfer function between r and the tracking error signal $r - y$. Note that all of the sensitivity functions are stable if the zeros of $\psi(z^{-1}, \rho)$ lie within the unit circle. Based on the internal model principle, the controller may contain the disturbance model to achieve a zero steady-state error. Therefore, the controllers may be pre-multiplied with any arbitrary function that actualizes the desired performance. For example, if it is desired to reject a step disturbance at the output, then $S(z^{-1}, \rho)$ can be pre-multiplied with $(1 - z^{-1})$ (which represents the integral action of the controller).

III. \mathcal{H}_∞ PERFORMANCE VIA CONVEX OPTIMIZATION

A. Preliminary Design Specifications

Suppose that it is desired to shape a particular sensitivity function. Consider, for example, the sensitivity function from d_o to y , expressed as $\mathcal{S}_0(z^{-1}, \rho)$. Given a stable weighting filter W_1 with a bounded infinity norm, a necessary and sufficient condition for achieving nominal performance is given by [14]:

$$\|W_1 \mathcal{S}_0\|_\infty < \gamma \quad (15)$$

where $\{\gamma \in \mathbb{R} \mid \gamma > 0\}$. In general, this condition can be extended to any of the sensitivity functions asserted in section (II-C). This condition can also be expressed as follows:

$$|W_1(e^{-j\omega})\mathcal{S}_0(e^{-j\omega}, \rho)| < \gamma; \quad \forall \omega \in \Omega \quad (16)$$

For notation purposes, the dependency in $e^{-j\omega}$ will be omitted, and will only be reiterated when deemed necessary. The dependency on ρ will continue to be highlighted. Notice that (16) can also be written as

$$\gamma^{-1}|W_1 MS(\rho)| < |\psi(\rho)|; \quad \forall \omega \in \Omega \quad (17)$$

where $\psi(\rho) = MS(\rho) + NR(\rho)$. It is desired to minimize the upper bound γ such that the \mathcal{H}_∞ performance condition is satisfied. Therefore, the following optimization problem can be considered:

$$\begin{aligned} & \underset{\{\rho_R, \rho_S, \rho_T\} \in \mathbb{R}}{\text{minimize}} && \gamma \\ & \text{subject to:} && \gamma^{-1}|W_1 MS(\rho)| < |\psi(\rho)| \\ & && \forall \omega \in \Omega \end{aligned} \quad (18)$$

Notice that (18) is both a non-linear and non-convex optimization problem.

Remark Note that the controller $T(z^{-1}, \rho)$ was not considered in the optimization problem (18) since $\mathcal{S}_0(z^{-1}, \rho)$ is independent of $T(z^{-1}, \rho)$. One of the benefits of implementing the RST controller is that the tracking performance can be specified independently from the regulation requirements, and thus the controller $T(z^{-1}, \rho)$ can be designed once $R(z^{-1}, \rho)$ and $S(z^{-1}, \rho)$ have been obtained. If tracking performance is crucial for the design, one may consider minimizing γ for $\|W_2 \mathcal{E}\|_\infty < \gamma$, where W_2 is also a stable transfer function with a bounded infinity norm.

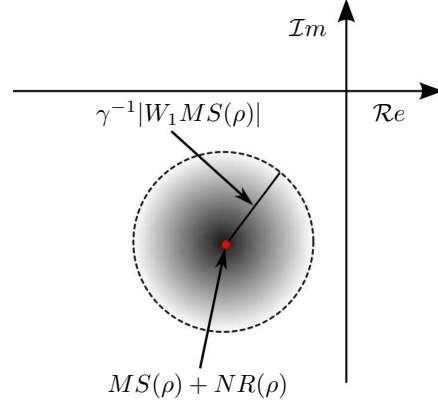


Fig. 2. The \mathcal{H}_∞ constraint in (15) can be represented as a circle in the complex plane. The constraint ensures that the circle will never encircle the origin for any frequency point in Ω .

Consider a circle in the complex plane which is centered at $MS(\rho) + NR(\rho)$ and has radius $\gamma^{-1}|W_1 MS(\rho)|$. The constraint in (17) ensures that for any frequency point in Ω , the circle associated with this frequency point will not encircle the origin. Figure 2 displays the graphical interpretation of this condition. According to [7], for each $\omega \in \Omega$ there exists a complex number $f(e^{-j\omega})$ which can rotate the disk in Fig. 2 such that it lays on the right hand side of the imaginary axis:

$$\text{Re}\{Z(\rho)f(e^{-j\omega})\} > 0; \quad \forall \omega \in \Omega \quad (19)$$

where $Z(\rho) = \psi(\rho) - \gamma^{-1}|W_1 MS(\rho)|$. In [15], it is shown that $f(e^{-j\omega})$ can be approximated arbitrarily well by the frequency response of a finite order transfer function if and only if

$$[\psi(\rho) - \gamma_0^{-1}|W_1 MS(\rho)|]^{-1}$$

is analytic in the right half plane for all $\gamma_0 > \gamma$. Based on (19) and on the results established in [7], the following theorem can be formulated:

Theorem 1 Given $\gamma > 0$, the performance criterion in (15) is satisfied if and only if there exist polynomials $R(z^{-1}, \rho)$ and $S(z^{-1}, \rho)$ such that

$$\gamma^{-1}|W_1 MS(\rho)| < \text{Re}\{\psi(\rho)\} \quad \forall \omega \in \Omega \quad (20)$$

where $\psi(\rho) = MS(\rho) + NR(\rho)$.

Proof: The proof of a similar theorem has been formally documented in [7]². ■

In certain design strategies, it may sometimes be desired to track different reference signals with no steady-state error, such as a step or a ramp input. For the torsional system that will be addressed in this paper, it will be desired to track a step input. Minimization of the error sensitivity function is a soft constraint, and may not lead to the ideal tracking

²This proof was established by assuming a unity feedback system with a rational controller structure $K = XY^{-1}$ in the feedforward loop. For the RST controller structure used in this paper, the controller R is equivalent to X and S is equivalent to Y .

performance. Therefore it is advantageous to consider conditions that ensure proper tracking of a step input by imposing hard convex constraints. Note that in an RST structure, the existence of an integrator in the open-loop transfer function does not guarantee a zero steady-state error for tracking a step input. The necessary and sufficient condition for a zero steady-state error is recalled in the following lemma.

Lemma 1 Suppose that the reference signal is a step function given as $r(z^{-1}) = A(1 - z^{-1})^{-1}$, where A is the amplitude of the step function. Additionally, suppose that the controller $S(z^{-1}, \rho)$ possesses an integrator (i.e., $S(z^{-1}, \rho) = (1 - z^{-1})S'(z^{-1}, \rho)$, where $S'(z^{-1}, \rho)$ is linearly parameterized). A necessary and sufficient condition to obtain a zero steady-state error for a step input is

$$R(1, \rho) = T(1, \rho) \neq 0 \quad (21)$$

Proof: The proof for this condition can be established by using the final value theorem. For perfect tracking of an arbitrary reference signal $r(k)$, it is required that $\lim_{k \rightarrow \infty} [r(k) - y(k)] = 0$, or

$$\lim_{z \rightarrow 1} (1 - z^{-1})r(z^{-1})[1 - \mathcal{T}(z^{-1}, \rho)] = 0 \quad (22)$$

For a step input, the condition for achieving a zero steady-state error can be expressed as

$$\lim_{z \rightarrow 1} [1 - \mathcal{T}(z^{-1}, \rho)] = 0 \quad (23)$$

By substituting (11) into (23) (and noting that $S(z^{-1}, \rho) = (1 - z^{-1})S'(z^{-1}, \rho)$), one can arrive to the following condition:

$$\begin{aligned} \lim_{z \rightarrow 1} \frac{N(z^{-1})[R(z^{-1}, \rho) - T(z^{-1}, \rho)]}{M(z^{-1})S(z^{-1}, \rho) + N(z^{-1})R(z^{-1}, \rho)} \\ = \frac{R(1, \rho) - T(1, \rho)}{R(1, \rho)} = 0 \end{aligned} \quad (24)$$

which evidently leads to the condition asserted in (21). ■

B. Convex Optimization via Semi-Definite Programming

Suppose that it is desired to achieve proper tracking performance and zero steady-state error for a step reference input. Moreover, it is desired that the sensitivity function $\mathcal{U}(z^{-1}, \rho)$ in (12) be shaped to limit the control efforts. As a result, the infinity norm of $W_2\mathcal{E}$ and $W_3\mathcal{U}$ should be minimized where W_2 is a lowpass filter with a cut off frequency close to the desired bandwidth and W_3 is a high pass filter to attenuate the control efforts at high frequencies. By utilizing the equality constraint in (21), one can formalize an optimization problem to obtain the admissible $R(\rho)$, $S(\rho)$, and $T(\rho)$ controllers as follows:

$$\begin{aligned} & \underset{\{\rho_R, \rho_S, \rho_T\} \in \mathbb{R}}{\text{minimize}} && \gamma \\ & \text{subject to:} && \gamma^{-1} |W_2[\psi(\rho) - NT(\rho)]| < \mathcal{R}e\{\psi(\rho)\} \\ & && \gamma^{-1} |W_3MT(\rho)| < \mathcal{R}e\{\psi(\rho)\} \\ & && R(1, \rho) = T(1, \rho) \neq 0 \\ & && \forall \omega \in \Omega \end{aligned} \quad (25)$$



Fig. 3. Torsional apparatus (ECP model 205a) used for the experimental analysis. The three disks are comprised of block masses which can be added or removed to alter the inertia of each disk (and thus alter the dynamics of the system). Each disk is vertically suspended on a spring with a variable spring constant. The actuator is located on the bottom of the device.

The optimization problem in (25) is quasi-convex since the objective function (γ) is being multiplied with the optimization parameters of $S(\rho)$ in one of the constraints. One solution is to implement an iterative bisection algorithm in order to convexify the optimization problem. In this method, an iterative approach is implemented in order to obtain an asymptotically convergent solution.

Remark In the bisection method, an initial value is assigned for γ such that $\gamma_i = 0.5(\gamma_{min} + \gamma_{max})$ to solve the optimization problem, where γ_{min} and γ_{max} are the minimum and maximum bounds set for γ . If the problem is feasible for γ_i , then $\gamma_{i+1} = 0.5(\gamma_{min} + \gamma_i)$, and the solution to the optimization problem in (25) is recalculated with γ_{i+1} . If the problem is infeasible for γ_i , then $\gamma_{i+1} = 0.5(\gamma_{max} + \gamma_i)$. This process is repeated until a solution is obtained within a given tolerance.

The problem in (25) is known as a semi-infinite programming (SIP) problem since there are a finite number of optimization variables and an infinite number of constraints. To solve this problem, the optimization algorithm can be converted to a semi-definite programming (SDP) problem. A predefined frequency space can be implemented in order to solve a finite number of constraints. This frequency grid can be predefined in a variety of manners (see [16], [17]).

IV. EXPERIMENTAL ANALYSIS

In order to demonstrate the effectiveness of the proposed method, an RST controller will be designed for a multi-model torsional apparatus (ECP Model 205a), as shown in Fig.3. This system contains three disks with variable inertias suspended vertically on anti-friction ball bearings. The disks are connected through a non-rigid cable with an adjustable spring constant. The actuator (a high torque brushless motor with a 2 Nm rating) is located at the bottom of the apparatus and is directly connected to the bottom disk via a rigid timing belt. The position of the disks are measured with a high resolution encoder (16,000 count/rev) and is used

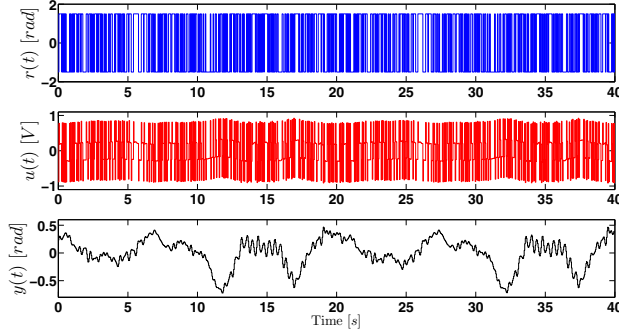


Fig. 4. Time-domain response of the closed-loop system with a PRBS excitation signal (shown only for the system configuration with two block masses on the top disk); the PRBS reference input $r(t)$ with a register length of 511 (solid-blue); control output $u(t)$ (solid-red); output response $y(t)$ (solid-black).

as feedback to control the closed-loop system. For this experiment, the dynamics of the apparatus will be altered by varying the inertias of the top disk.

An RST controller will be designed for various inertial configurations. Three different configurations will be considered for this design; the bottom disk and the middle disk will possess fixed inertias, while the inertia for the top disk will be varied. The inertia is varied by altering the number of block masses that are arranged on the disk.

The input to the system is the current of the actuator and the output is the position of the third disk. Therefore, the plant model has an integrator and is not stable; thus it is required to obtain the FRF's of the various configurations in a closed-loop structure. As asserted in Section (II-A), a stabilizing controller must be used to obtain $N(e^{-j\omega})$ and $M(e^{-j\omega})$. For this plant, a PID controller was designed to stabilize the closed-loop system. A reference input with a pseudo-random binary sequence (PRBS) is implemented to excite the closed-loop system with a sampling frequency of 25 Hz. The time-domain signals for the reference input $r(t)$, control output $u(t)$, and output $y(t)$ for this system are depicted in Fig.4 (for brevity, the figure shown is for one of the configurations). The FRF of $N(e^{-j\omega})$ is then obtained with the spectral analysis command in MATLAB (i.e., `spa(·)`) by using the time-domain data of $r(t)$ and $y(t)$. Similarly, the FRF of $M(e^{-j\omega})$ is obtained in a similar fashion by using the time-domain data of $r(t)$ and $u(t)$. The FRF of the plant model can then be described as $G(e^{-j\omega}) = N(e^{-j\omega})M^{-1}(e^{-j\omega})$. The FRF's for each of the three configurations are shown in Fig. 5.

It will be desired to minimize the tracking error and to reject a step disturbance at the output. In order to achieve these specifications, an integrator will be included in $S^*(z^{-1}, \rho)$ (i.e., $S^*(z^{-1}, \rho) = (1 - z^{-1})S(z^{-1}, \rho)$). Additionally, a weight will be designated to limit the control effort. Therefore, the optimization problem formulated in (25) will be implemented for this design scheme. The RST controllers will each be linearly parameterized as in (3), (4), (5) with $n_r = n_s = n_t = 8$. Thus the following optimization

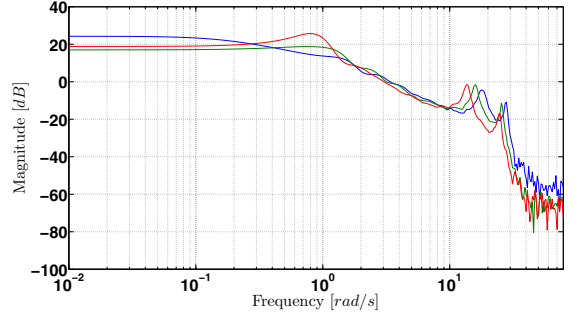


Fig. 5. FRF's of the plant model obtained from the closed-loop time-domain response of each system configuration. The loads on the bottom and middle disk are fixed while the load on the top disk is varied: FRF with one block mass on the top disk (solid-blue); FRF with two block masses on the top disk (solid-green); FRF with four block masses on the top disk (solid-red).

problem will be considered:

$$\begin{aligned} & \text{minimize} \quad \gamma \\ & \{ \rho_R, \rho_S, \rho_T \} \in \mathbb{R} \\ & \text{subject to:} \\ & \gamma^{-1} |W_2(e^{-j\omega_k})[\psi_i^*(e^{-j\omega_k}, \rho) - N_i(e^{-j\omega_k})T(e^{-j\omega_k}, \rho)]| \\ & \quad - \mathcal{R}e\{\psi_i^*(e^{-j\omega_k}, \rho)\} < 0 \\ & \gamma^{-1} |W_3(e^{-j\omega_k})M_i(e^{-j\omega_k})T(e^{-j\omega_k}, \rho)| \\ & \quad - \mathcal{R}e\{\psi_i^*(e^{-j\omega_k}, \rho)\} < 0 \\ & R(1, \rho) = T(1, \rho) \neq 0 \\ & \text{for } i = 1, 2, 3 \text{ and } k = 1, \dots, q \end{aligned} \quad (26)$$

where

$$\begin{aligned} \psi_i^*(e^{-j\omega_k}, \rho) &= M_i(e^{-j\omega_k})S^*(e^{-j\omega_k}, \rho) \\ & \quad + N_i(e^{-j\omega_k})R(e^{-j\omega_k}, \rho) \end{aligned}$$

A. Weighting filter selection

The weighting filter W_2 will be chosen based on the fact that $\mathcal{E}_d + \mathcal{T}_d = 1$, where \mathcal{E}_d is the desired error sensitivity function and \mathcal{T}_d is the desired complementary sensitivity function (i.e., the closed-loop transfer function). \mathcal{T}_d is chosen as a first order transfer function such that the step response will have a time constant of 1 s (which corresponds to a closed-loop bandwidth of $\omega_b = 1 \text{ rad/s}$). The transfer function which satisfies these requirements can be formulated as

$$\mathcal{T}_d(s) = \frac{\omega_b}{s + \omega_b} \quad (27)$$

Since $\mathcal{E}_d = 1 - \mathcal{T}_d$, an appropriate filter for the error sensitivity function can be devised as

$$W_2(s) = \frac{s + \omega_b}{s} \quad (28)$$

It will be desired to minimize the control effort at high frequencies; therefore, a viable choice for the control weighting function W_3 is

$$W_3(s) = \frac{s + \omega_u/M_u}{\omega_u} \quad (29)$$

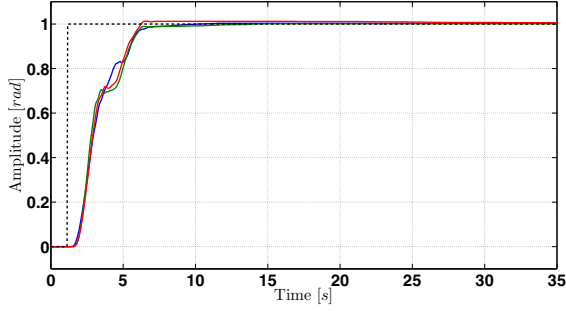


Fig. 6. Step response for each load configuration: response with one block mass on the top disk (solid-blue); response with two block masses on the top disk (solid-green); response with four block masses on the top disk (solid-red).

where M_u is the maximum controller gain, and ω_u is the controller bandwidth [18]. For the torsional system considered in this paper, an appropriate choice for these parameters are $M_u = 10^3$ and $\omega_u = 3 \text{ rad/s}$.

B. Experimental results

The optimization problem in (26) is solved by considering a logarithmically spaced frequency grid with $q = 400$ points. The SDPT3 software package is used in conjunction with Matlab to solve the optimization problem. The solution obtained from the bisection algorithm produces the following controllers:

$$\begin{aligned} R(z^{-1}) &= 34.19 - 113.7z^{-1} + 160z^{-2} \\ &\quad - 100.3z^{-3} - 10.66z^{-4} + 60.08z^{-5} \\ &\quad - 32.94z^{-6} - 2.093z^{-7} + 5.542z^{-8} \\ S^*(z^{-1}) &= 1 + 0.3538z^{-1} + 0.2304z^{-2} - 0.07627z^{-3} \\ &\quad - 0.2845z^{-4} - 0.2575z^{-5} - 0.08136z^{-6} \\ &\quad - 0.1705z^{-7} - 0.2921z^{-8} - 0.422z^{-9} \\ T(z^{-1}) &= 0.00942 + 0.01295z^{-1} + 0.01548z^{-2} \\ &\quad + 0.01646z^{-3} + 0.01116z^{-4} + 0.01683z^{-5} \\ &\quad + 0.01491z^{-6} + 0.01329z^{-7} + 0.01008z^{-8} \end{aligned}$$

The optimal value of γ obtained from the optimization is $\gamma_{opt} = 2.228$. The step responses obtained for each of the inertial configurations are depicted in Fig. 6. From the multi-model step responses, it can be observed that the system is stable and robust to the dynamic variations of the torsional apparatus. Moreover, the load variations do not significantly impact the tracking performance (which is expected, since the same weighting filter was used in the multi-model optimization problem). The closed-loop FRF's of all three system configurations are shown in Fig. 7. It can be observed that the achieved closed-loop bandwidth for all three configurations is approximately 1 rad/s , which is the bandwidth that was specified to form the weighting filter $W_2(s)$. This confirms the feasibility of the solution obtained from the optimization problem in (26).

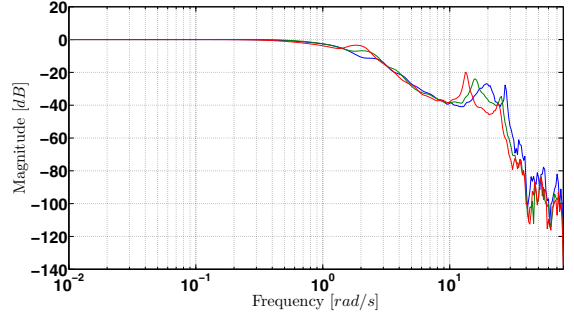


Fig. 7. Closed-loop frequency response functions of all three system configurations: closed-loop FRF with one block mass on the top disk (solid-blue); closed-loop FRF with two block masses on the top disk (solid-green); closed-loop FRF with four block masses on the top disk (solid-red).

V. CONCLUSION

In this paper, a new method has been proposed to design robust controllers with \mathcal{H}_∞ performance. The RST controller structure possesses many practical advantages, such as its two-degree of freedom design capabilities and the fact that it can be easily implemented. A frequency-domain approach has been used in order to avoid the problem of unmodeled dynamics associated with parametric models. A convex optimization problem is constructed based on the \mathcal{H}_∞ criterion thanks to the process of linearly parameterizing the RST controllers. This controller design method has been employed to design a robust RST controller to control the disk position of a coupled torsional apparatus. A multi-model optimization approach was considered to design a controller for various loads; the design proved to be robust to the dynamic variations of the system. For further research, it will be desired to establish an optimal frequency gridding process, and compare how different gridding schemes can improve the solution to the optimization problem.

REFERENCES

- [1] S. Khadraoui, H. Nounou, M. Nounou, A. Datta, and S. Bhattacharyya, "A control design method for unknown systems using frequency domain data," in *Control Conference (ASCC), 2013 9th Asian*. IEEE, 2013, pp. 1–6.
- [2] M. Saeki, "Data-driven loop-shaping design of PID controllers for stable plants," *International Journal of Adaptive Control and Signal Processing*, 2013.
- [3] S. Khadraoui, H. Nounou, M. Nounou, A. Datta, and S. Bhattacharyya, "Robust control design method for uncertain system using a set of measurements," in *American Control Conference (ACC), 2013*. IEEE, 2013, pp. 4325–4330.
- [4] K. Yubai, D. Yashiro, J. Hirai *et al.*, "Design of a PID controller based on H_∞ loop shaping method using frequency responses," in *Control, Automation and Systems (ICCAS), 2013 13th International Conference on*. IEEE, 2013, pp. 966–969.
- [5] A. Karimi and G. Galdos, "Fixed-order H_∞ controller design for nonparametric models by convex optimization," *Automatica*, vol. 46, no. 8, pp. 1388–1394, 2010.
- [6] A. Karimi, G. Galdos, and R. Longchamp, "Robust fixed-order H_∞ controller design for spectral models by convex optimization," in *47th IEEE Conference on Decision and Control*, Cancun, MEX, 2008.
- [7] A. Karimi and Y. Zhu, "Robust H-infinity Controller Design Using Frequency-Domain Data," in *19th IFAC World Congress*, 2014.
- [8] I. D. Landau and G. Zito, *Digital control systems: design, identification and implementation*. Springer, 2006.

- [9] M. Bendjedia, K. Tehrani, and Y. Azzouz, "Design of rst and fractional order pid controllers for an induction motor drive for electric vehicle application," 2014.
- [10] M. Larbi, S. Hassaine, S. Moreau, and B. Mazari, "Speed control by rst with load observer of a permanent magnet synchronous motor," in *Power Electronics and Motion Control Conference (EPE/PEMC), 2010 14th International*. IEEE, 2010, pp. T10–20.
- [11] M. Khanchoul and M. Hilairret, "Design and comparison of different rst controllers for pmsm control," in *IECON 2011-37th Annual Conference on IEEE Industrial Electronics Society*. IEEE, 2011, pp. 1795–1800.
- [12] R. Madiouni, S. Bouallegue, J. Haggege, and P. Siarry, "Particle swarm optimization-based design of polynomial rst controllers," in *Systems, Signals & Devices (SSD), 2013 10th International Multi-Conference on*. IEEE, 2013, pp. 1–7.
- [13] G. Galdos, A. Karimi, and R. Longchamp, "RST controller design by convex optimization using frequency-domain data," in *Proceedings of the 18th IFAC World Congress*, Milan, Italy, 2011.
- [14] K. Zhou, *Essentials of Robust Control*. New Jersey: Prentice Hall, 1998.
- [15] A. Rantzer and A. Megretski, "Convex parameterization of robustly stabilizing controllers," *IEEE Trans. on Automatic Control*, vol. 39, no. 9, pp. 1802–1808, September 1994.
- [16] P. Seiler, B. Vanek, J. Bokor, and G. J. Balas, "Robust H_∞ filter design using frequency gridding," in *American Control Conference (ACC), 2011*. IEEE, 2011, pp. 1801–1806.
- [17] G. Galdos, A. Karimi, and R. Longchamp, " H_∞ controller design for spectral MIMO models by convex optimization," *Journal of Process Control*, vol. 20, no. 10, pp. 1175 – 1182, 2010.
- [18] K. Zhou and J. C. Doyle, *Essentials of robust control*. N.Y.: Prentice-Hall, 1998.